

K-Nearest Neighbors

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High values of K will result in smoother decision boundaries
 \implies lower variance but also higher bias

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What is the time complexity of finding k nearest neighbors naively?

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- ▶ **Scalability:** Approximate methods needed for large datasets